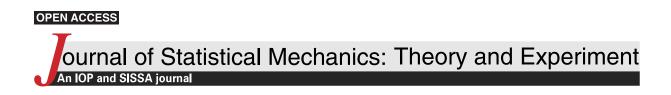


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Geometrical optics of large deviations of Brownian motion in inhomogeneous media

To cite this article: Tal Bar and Baruch Meerson J. Stat. Mech. (2023) 093301

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PAPER: Disordered systems, classical and quantum

Geometrical optics of large deviations of Brownian motion in inhomogeneous media

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Received 15 May 2023 Accepted for publication 4 August 2023 Published 12 September 2023

Online at stacks.iop.org/JSTAT/2023/093301 https://doi.org/10.1088/1742-5468/acf125

Abstract. Geometrical optics provides an instructive insight into Brownian motion, 'pushed' into a large-deviations regime by imposed constraints. Here we extend geometrical optics of Brownian motion by accounting for diffusion inhomogeneity in space. We consider three simple model problems of Brownian motion on the line or in the plane in situations where the diffusivity of the Brownian particle depends on one spatial coordinate. One of our results describes 'Brownian refraction': an analog of refraction of light passing through a boundary between two media with different refraction indices.

Keywords: Brownian motion [100], large deviations in non-equilibrium systems [100], fluctuation phenomena [75]

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1. Introduction

Large deviations of Brownian motion have recently attracted much attention from statistical physicists [1–13]. One setting where large deviations arise naturally is stochastic searching of a distant target by multiple random searchers, such as receptors searching for a reaction site on the surface of a living cell [14], or sperm cells searching for an oocyte [15]. In such systems the multiple searchers essentially compete among themselves for reaching the target first and performing a biological function. When the number of searchers is very large (for example, there are about 3×10^8 sperm cells which attempt to reach the oocyte after copulation in humans), the searcher which succeeds in arriving first does it in an unusually short time. As a result, this searcher's trajectory looks quite differently from *typical* trajectories of a single searcher conditioned on reaching the target at any time [1–3, 5–9, 11, 12].

For a single Brownian searcher, the probability of an extremely fast arrival at the target is very small; it corresponds to a short-time tail of the complete probability distribution of the arrival times. Short-time tails of this type can be efficiently described theoretically by the optimal fluctuation method (OFM) which, in the context of Brownian motion, is called 'geometrical optics of Brownian motion' [1-3, 5-9, 11, 12]. An important advantage of geometrical optics is that it allows to account, in a simple manner, for additional constraints. Furthermore, apart from providing the distribution tail of interest, geometrical optics predicts the optimal path of the system, that is the most likely trajectory of the Brownian particle which dominates the probability distribution tail. The optimal path gives a valuable insight into the physics of the large deviation in question.

Previous works on geometrical optics of Brownian motion assumed that the Brownian motion, 'pushed' into a large-deviation regime by imposed constraints, occurs in a homogeneous medium (possibly with some impenetrable regions) [1–3, 5–9, 11, 12]. In some applications, however, the homogeneity assumption is unrealistic. For example, the diffusivity of a reactant macromolecule, exploring a living cell can vary quite significantly due to a varying physical and chemical composition of different cell regions [16]. Here we consider large deviations of a Brownian particle with a space-dependent diffusivity, as described by the Langevin equation

$$\dot{\mathbf{x}} = \sqrt{2D[\mathbf{x}(t)]} \boldsymbol{\xi}(t) \,. \tag{1}$$

Here $\mathbf{x}(t)$ is the displacement of the particle, $D = D(\mathbf{x})$ is the space-dependent diffusivity, and $\boldsymbol{\xi}(t)$ is the Gaussian white noise with unit norm and zero mean. Using the Hänggi–Klimontovich interpretation of the multiplicative noise in equation (1) [17], one can write down the diffusion equation

$$\frac{\partial P(\mathbf{x},t)}{\partial t} = \nabla \cdot [D(\mathbf{x})\nabla P(\mathbf{x},t)]$$
(2)

for the probability $P(\mathbf{x},t)$ of observing the particle at position \mathbf{x} at time t. In one spatial dimension equations (1) and (2) were studied in [18] in connection to phenomena accompanying anomalous diffusion, and in [19] in the context of fluctuation-dissipation relation. Here we are interested in large deviations in this system. The starting point of our analysis will be the Wiener's action

$$-\ln P[\mathbf{x}(t)] \simeq S = \int_0^T \frac{\dot{\mathbf{x}}^2(t)}{4D[\mathbf{x}(t)]} dt, \qquad (3)$$

which gives the probability $\mathcal{P}[\mathbf{x}(t)]$ of a Brownian path $\mathbf{x}(t)$ up to pre-exponential factors. A large-deviation regime corresponds to a very large action, $S \gg 1$. It is this regime where the (tail of the) probability distribution of interest is dominated by a single optimal path $\mathbf{x}(t)$: a deterministic trajectory which minimizes the Wiener's action (3) over all possible trajectories subject to additional, problem-specific constraints.

This is the essence of the method of geometrical optics of large deviations of Brownian motion [1–3, 5–9, 11, 12], previously developed for homogeneous media. For D = const (and in the absence of any integral constraints) the action (3) is minimized by a particle motion with constant speed, $|\dot{\mathbf{x}}| = \text{const}$, along a geodesic: the shortest path between the initial and final position of the particle, subject to additional constraints, if any. The action along such a path is

$$S = \frac{1}{4D} \int_0^T \left(\frac{\mathcal{L}}{T}\right)^2 dt = \frac{\mathcal{L}^2}{4DT},\tag{4}$$

where \mathcal{L} is the path's length. For D = const the problem therefore reduces to minimizing \mathcal{L} subject to the additional constraints [7].

Here we are interested in inhomogeneous diffusion as described by equations (1)-(3). To this end we consider three simple model problems where D depends only on one spatial coordinate x. As we will see, the diffusion inhomogeneity can lead to interesting effects. One of them is the 'Brownian refraction', which emphasizes the geometrical nature of large deviations of Brownian motion and provides an additional justification to the term 'geometrical optics of diffusion'.

Here is a layout of the remainder of the paper. In section 2 we use the geometrical optics to determine the short-time asymptotic of the propagator of the Brownian motion on the line with a piecewise constant diffusivity D(x). In this case the propagator can be also found exactly by solving equation (2) (see the appendix). This allows us to verify geometrical optics' predictions. Importantly, the latter also include the optimal path of the system, which is not directly available from the exact solution. In section 3 we find the short-time asymptotic of the propagator of the Brownian motion on the plane with a piecewise constant D(x). It is here where we encounter the Brownian refraction. In section 4 the geometrical optics is used to determine the short-time asymptotic of the propagator of the Brownian motion on the line with a continuously varying diffusivity D(x). A brief discussion of our results and of their possible extensions is included in section 5.

2. Brownian motion on the line with a piecewise constant D(x)

Our first example is Brownian motion on a line with a piecewise constant diffusivity:

$$D(x) = \begin{cases} D_1, & x < \ell, \\ D_2 \equiv \kappa^2 D_1, & x > \ell, \end{cases}$$

$$(5)$$

where $\ell > 0$. The particle starts at the origin at t = 0, and we are interested in the shorttime, $T \to 0$, asymptotic of the probability density P(L,T) of observing the particle at a point $x = L > \ell$ at time T > 0.

On each of the intervals $0 < x < \ell$ and $\ell < x < L$ the diffusivity is constant. Therefore, in view of the above-mentioned geometrical optics results for D = const, the optimal path x(t) must be a piecewise linear function of time. Let us denote by τ the (*a priori* unknown) time of crossing the boundary $x = \ell$ between the two constant-D regions (we assume that this time is unique.) Demanding that x(t) obey the conditions x(0) = 0and x(T) = L, we obtain

$$x(t) = \begin{cases} \frac{\ell t}{\tau}, & 0 < t < \tau, \\ \frac{\ell (T-t) + L(t-\tau)}{T-\tau}, & \tau < t < T. \end{cases}$$
(6)

The action on this composite trajectory is the sum of contributions, described by equation (4), from each of the two intervals:

$$S \simeq \frac{\ell^2}{4D_1\tau} + \frac{(L-\ell)^2}{4D_2(T-\tau)} = \frac{1}{4} \left(\frac{t_1}{\tau} + \frac{t_2}{T-\tau} \right), \tag{7}$$

where

$$t_1 = \frac{\ell^2}{D_1}$$
 and $t_2 = \frac{(L-\ell)^2}{D_2} = \frac{(L-\ell)^2}{\kappa\sqrt{D_1}}$ (8)

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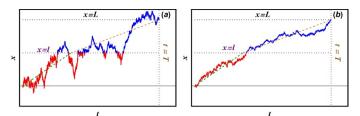


Figure 1. Solid lines: two simulated Brownian paths for the setting of section 2. The parameters are L = 6, $\ell = 3$, $D_1 = 10$, $D_2 = 3$, and T = 6 (a) and T = 0.45 (b). The dashed lines show the piecewise linear optimal path predicted by geometrical optics, equations (6) and (9).

are the characteristic diffusion times through each of the two media. The minimum action is achieved for

$$\tau = \frac{\sqrt{t_1}}{\sqrt{t_1} + \sqrt{t_2}} T = \frac{T}{\frac{1}{\kappa} \left(\frac{L}{\ell} - 1\right) + 1},\tag{9}$$

and we obtain the short-time asymptotic

$$-\ln P(L, T \to 0) \simeq S_{\min} = \frac{\left(\sqrt{t_1} + \sqrt{t_2}\right)^2}{4T} = \frac{\left[L - (1 - \kappa)\ell\right]^2}{4D_2T}.$$
(10)

The same $T \to 0$ asymptotic (10) follows from exact solution of this problem (see the appendix for details):

$$P_{\text{exact}}(L,T) = \frac{1}{\sqrt{\pi T} \left(\sqrt{D_1} + \sqrt{D_2}\right)} e^{-\frac{[L-(1-\kappa)\ell]^2}{4D_2T}}.$$
(11)

Indeed, when $T \to 0$, the pre-exponential factor in equation (11) (which is missed by the leading-order geometrical optics calculations) becomes subleading, and equations (10) and (11) coincide in the leading order.

Figures 1(a) and (b) gives two examples of simulated Brownian paths in this setting. The particle starts at the origin and arrives at x = L at two different times T. The figure also shows the piecewise linear optimal path x(t) at $T \to 0$, predicted by equations (6) and (9) for the same D_1 and D_2 . As T decreases, the simulated paths start approaching the predicted short-time optimal path.

Notice that the geometrical optics result (10) gives a correct leading-order asymptotic $P(L, T \to 0)$ for a whole class of settings dealing with a Brownian particle starting at the origin and arriving at x = L at t = T. One such setting—when the Brownian motion is defined on the whole line—is solved exactly in the appendix. But one could actually consider a Brownian motion on any interval (a, b) which includes in itself the interval 0 < x < L, with either reflecting or absorbing boundary conditions at the boundaries x = a and x = b. In this case the exact solution would of course be different from equation (11), but its leading-order $T \to 0$ asymptotic will still be described by

equation (10). This is explained by the character of the optimal path: when conditioned on reaching x = L in a very short time, the particle cannot spend any time on exploring regions beyond the interval (0, L), rendering them irrelevant for the leading-order calculations.

3. Brownian motion on the plane with a piecewise constant D(x): Brownian refraction

Now let us consider a Brownian motion in the (x, y) plane with a piecewise constant diffusivity

$$D(x) = \begin{cases} D_1, & x < \ell, \\ D_2 \equiv \kappa^2 D_1, & x > \ell, \end{cases}$$
(12)

for any y. The Brownian particle starts at the origin at t=0, and we are interested in the short-time asymptotic, $T \to 0$, of the probability $P(L_x, L_y, T)$ for the particle to reach the point (L_x, L_y) , so that $L_x > \ell$, at time t = T, see figure 2.

We assume that the particle crosses the boundary $x = \ell$ between the two regions with different diffusivities at a unique time $0 < \tau < T$, and the *y*-coordinate of the particle at this moment is y_0 . The action on this trajectory is the sum of two terms, described by equation (4), which come from each of the two media:

$$S(\tau, y_0) = \frac{1}{4} \left[\frac{\ell^2 + y_0^2}{D_1 \tau} + \frac{(L_x - \ell)^2 + (L_y - y_0)^2}{D_2 (T - \tau)} \right] = \frac{1}{4} \left(\frac{t_1}{\tau} + \frac{t_2}{T - \tau} \right), \quad (13)$$

with obvious definitions of the characteristic diffusion times t_1 and t_2 for each of the two media. Now we should minimize the action (13) with respect to the two auxiliary parameters τ and y_0 . It is convenient to do it sequentially. Notice that equations (7) and (13) have the same form in terms of τ , t_1 and t_2 . Therefore, at fixed y_0 , the minimum of the action with respect to τ is achieved at $\tau = \frac{\sqrt{t_1}}{\sqrt{t_1} + \sqrt{t_2}}T$ (see equation (9)), and we obtain

$$S(y_0) = \frac{1}{4T} \left(\sqrt{t_1} + \sqrt{t_2}\right)^2 = \frac{1}{4T} \left[\sqrt{\frac{\ell^2 + y_0^2}{D_1}} + \sqrt{\frac{\left(L_x - \ell\right)^2 + \left(L_y - y_0\right)^2}{D_2}} \right]^2.$$
(14)

The minimization with respect to the remaining parameter y_0 demands $dS(y_0)/dy = 0$, which leads to the relation

$$\frac{L_y - y_0}{\sqrt{D_2}\sqrt{(L_x - l)^2 + (L_y - y_0)^2}} = \frac{y_0}{\sqrt{D_1}\sqrt{\ell^2 + y_0^2}}.$$
(15)

To better understand this result we notice that y_0 and $L_y - y_0$ can be written as $y_0 = \ell \tan \theta_1$, and $L_y - y_0 = (L_x - \ell) \tan \theta_2$, respectively, where θ_1 (θ_2) is the angle between the

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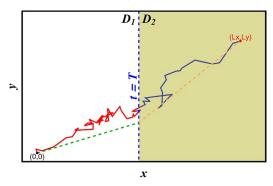


Figure 2. A simulated Brownian path for the setting of section 3 with $\kappa > 1$. The parameters are $L_x = L_y = 10$, $\ell = 5$, $D_1 = 1$, $D_2 = 3$, and T = 4.17 (solid line). The dashed lines show the piecewise linear optimal path predicted by the geometrical optics in the limit of $T \rightarrow 0$. This path obeys the Snell's law (16).

particle trajectory in the left (right) region and the straight line $y = y_0$. In other words, θ_1 and θ_2 are the angles of incidence and refraction, respectively. In the new notation equation (15) becomes, remarkably,

$$\frac{\sin\theta_1}{\sin\theta_2} = \frac{\sqrt{D_1}}{\sqrt{D_2}} = \frac{1}{\kappa},\tag{16}$$

providing an unexpected Brownian-motion analog of Snell's law of optics (the latter describes the refraction of light or other waves passing through a boundary between two media with different refraction indices [20, 21]). Notice that the role of the refraction index n in optics is played here by $1/\sqrt{D}$. The 'Brownian refraction', described by equation (16), provides a further justification to the term 'geometrical optics of (large deviations of) Brownian motion' [1–3, 5–9, 11, 12]¹.

Let us complete the solution. The relation (15) yields a quartic equation for y_0 , the analytical solutions of which is too bulky to work with. It is more convenient to present the solution in a parametric form. Using equation (15), we can express L_y via y_0 :

$$L_y = y_0 \left[1 + \frac{(L_x - l)\kappa}{\sqrt{l^2 + (1 - \kappa^2)y_0^2}} \right].$$
 (17)

Plugging this expression into equation (14), we obtain

$$S = \frac{\ell^2 + y_0^2}{4D_2T} \left[\kappa + \frac{(L_x - \ell)}{\sqrt{\ell^2 + (1 - \kappa^2)y_0^2}} \right]^2.$$
(18)

¹ To remind the reader, the Snell's law of refraction can be derived by minimizing the arrival time of the ray to the target point [21]. For the Brownian motion the arrival time is fixed, and one minimizes the Wiener's action (3). In the light of these differences the Brownian refraction phenomenon is somewhat surprising.

Equations (17) and (18) yield $-\ln P \simeq S$ (which we are after) in a parametric form, where y_0 serves as the parameter.

It is instructive to look at the limits of S at $\kappa \to 0$ and $\kappa \to \infty$ at fixed D_2 :

$$S(\kappa) \simeq \begin{cases} \frac{(L_x - \ell)^2}{4D_2 T}, & \kappa \to 0, \\ \frac{\ell^2}{4D_1 T}, & \kappa \to \infty. \end{cases}$$
(19)

The corresponding optimal paths in these two limits are the following. At $\kappa \to 0$ we have $y_0 \to L_y$, and $\tau \to 0$. The form of the trajectory is

$$y(x, \kappa \to 0) \simeq \begin{cases} \frac{L_y}{\ell} x, & 0 < x < \ell, \\ L_y, & \ell < x < L_x. \end{cases}$$
(20)

Here the diffusion in media 1 is fast, and the particle reaches the point (ℓ, L_y) almost immediately so as to minimize the distance it has to travel in media 2, where the diffusivity is finite. The particle spends almost all of the allocated time T in media 2.

As $\kappa \to \infty$, y_0 goes to zero, and τ approaches T. Now the form of the trajectory is

$$y(x, \kappa \to \infty) \simeq \begin{cases} 0, & 0 < x < \ell, \\ \frac{L_y(x-\ell)}{L_x-\ell}, & \ell < x < L_x. \end{cases}$$
(21)

Here the diffusion in media 1 is slow. Therefore, the particle moves along the x-axis so as to minimize the distance it has to travel here, and its spends almost all of the allocated time T to arrive at $x = \ell$. Then it almost immediately 'jumps' to the point (L_x, L_y) along the straight line.

4. Brownian motion on the line with a continuously varying D(x)

Here we return to one spatial dimension and suppose that a Brownian particle is released at t = 0 at x = 0 in a medium with a continuously varying diffusivity D(x).

We are interested in the short-time asymptotic of the probability P(L,T) of observing the particle at x = L at time t = T. Using geometrical optics, we identify the Lagrangian

$$\mathfrak{L}(x,\dot{x}) = \frac{\dot{x}^2}{4D(x)} \tag{22}$$

of the Wiener's action (3). Since this Lagrangian does not depend explicitly on time, there is an integral of motion which, in this case, coincides with the Lagrangian itself, $\mathfrak{L}(x, \dot{x}) = \text{const.}$ Therefore, we can write

$$\dot{x} = C\sqrt{D(x)},\tag{23}$$

where C is a constant to be determined. A comparison of equation (23) with the Langevin equation (1) shows that the optimal realization of the rescaled Gaussian white noise $\xi(t)$, conditioned on a very fast arrival of the Brownian particle at x = L, is simply a constant, equal to $C/\sqrt{2}$.

Equation (23) can be solved in quadratures for any D(x), and we obtain

$$Ct = \int_0^x \frac{dy}{\sqrt{D(y)}},\tag{24}$$

where we used the boundary condition x(0) = 0. Using the second boundary condition x(T) = L, we determine the constant C:

$$C = \frac{1}{T} \int_0^L \frac{dx}{\sqrt{D(x)}}.$$
(25)

Equations (24) and (25) yield the optimal path x(t) in an implicit form for any D(x). Further, plugging equation (23) into equation (3) and using equation (25), we obtain a general expression for the action:

$$S = \int_0^T \frac{\dot{x}^2}{4D(x)} dt = \frac{1}{4T} \left[\int_0^L \frac{dx}{\sqrt{D(x)}} \right]^2.$$
 (26)

As a simple example, consider a linear diffusivity profile, $D(x) = D_0(1 + \beta x/L)$, where we must assume $\beta > -1$ in order to keep D(x) positive in the relevant region 0 < x < L. Using equations (24), we obtain

$$x(t) = L\frac{t}{T} \left[\frac{t}{T} + \frac{2\left(\sqrt{\beta+1}-1\right)}{\beta} \left(1-\frac{t}{T}\right) \right].$$
(27)

The action (26) takes the form

$$S = \frac{L^2}{D_0 T} f(\beta), \text{ where } f(\beta) = \frac{\left(\sqrt{\beta + 1} - 1\right)^2}{\beta^2}.$$
 (28)

It gives the short-time asymptotic of P(L,T) via the relation $-\ln P(L,T) \simeq S$. A graph of the function $f(\beta)$ is shown in figure 3. As one can see, the action decreases monotonically with an increase of β . It reaches its maximum value $L^2/(DT)$ at $\beta = -1$. In this case the diffusivity vanishes at x = L. As β goes to infinity, $f(\beta)$ goes down as $1/\beta$. Finally, at $\beta \to 0$ we reproduce the constant-diffusivity action $S = L^2/(4D_0T)$.

The optimal path x(t), as described by equation (27), is shown in figure 4 for three different values of β . In general, x(t) is a parabola: concave or convex depending on whether β is positive or negative, respectively. For $\beta = 0$ (a constant diffusivity) x(t) is a straight line as to be expected.



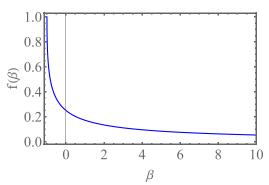


Figure 3. The function $f(\beta)$ from equation (28).

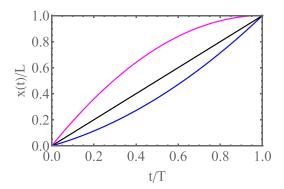


Figure 4. The optimal path x(t), as described by equation (27), for the linear diffusivity profile $D(x) = D_0(1 + \beta x/L)$ with $\beta = 10$ (blue) and -1 (magenta). The black line shows x(t) for the constant diffusivity, $\beta = 0$.

5. Discussion

We considered three simple model problems, illustrating the role of spatial variations of the diffusivity in large deviations of Brownian motion. One of our predictions is what can be called 'Brownian refraction': a close analog of refraction of light passing through a boundary between two media with different refraction indices. This phenomenon appears when one conditions the Brownian particle on arrival at a distant point in a very short time.

It would be interesting to extend the geometrical-optics approach by accommodating additional local or integral constraints, or constraints in the form of inequalities, like it was done for the homogeneous diffusion [1-3, 5-9, 11, 12].

When dealing, in section 3, with a Brownian motion on the plane, we assumed that the diffusivity D depends only on one spatial coordinate x, and this dependence is piecewise constant. This theory can be extended to a continuously varying D(x). Indeed, the Lagrangian of equation (3),

$$\mathfrak{L}(x, \dot{x}, \dot{y}) = \frac{\dot{x}^2}{4D(x)} + \frac{\dot{y}^2}{4D(x)}$$
(29)

describes an effective classical mechanics with two degrees of freedom. This system possesses two integrals of motion. The first of them is the Lagrangian itself,

$$\frac{\dot{x}^2}{4D(x)} + \frac{\dot{y}^2}{4D(x)} = E = \text{const.}$$
(30)

An additional integral of motion, $\dot{y}/D[x(t)] = \text{const}$, follows from the fact that the Lagrangian (29) does not depend on y. The presence of two integrals of motion allows one to solve the problem in quadratures for any D(x).

Extensions of the theory to a more general case, where D(x,y) depends both on x, and on y, encounter an immediate difficulty [22], because in this case the geometricaloptics description of the system involves only one integral of motion: the Lagrangian $\mathfrak{L}(x,y,\dot{x},\dot{y})$ itself. The lack of additional independent integral of motion makes the geometrical-optics formulation non-integrable analytically [23]. Numerical solutions, however, are certainly possible. In addition, perturbative analytical treatments can be possible if there is a small parameter in the coordinate dependence D(x,y). It would be interesting to explore these possibilities in a future work.

Finally, geometrical optics of large deviations of Brownian motions present only one, although important, example of a class of systems which can be studied with the OFM. In recent years the OFM has been employed for analysis of large deviations in additional Markovian [24] and non-Markovian [25–28] stochastic processes. Another well-known large-deviation technique, based on the Feynman–Kac formula [29, 30], applies to a certain class of large deviations of *time-averaged* quantities. There is also a group of large-deviation methods, in different areas of physics, which rely on the 'single-bigjump' principle, see [31] and references therein. A skillful use of all these methods (and sometimes of their combinations [32]) should help develop a better understanding of many important, fascinating and, quite often, non-intuitive large deviations and rare events.

Acknowledgments

We thank Oded Farago for useful comments. B M was supported by the Israel Science Foundation (Grant No. 1499/20).

Appendix. Exact solution for section 2

Here we present exact solution of the equation

$$\frac{\partial P(x,t)}{\partial t} = \frac{\partial}{\partial x} \left[D(x) \frac{\partial P}{\partial x} \right]$$
(A1)

for the probability density P(x,t) on an infinite line $|x| < \infty$, subject to the initial condition

$$P(x,t=0) = \delta(x). \tag{A2}$$

https://doi.org/10.1088/1742-5468/acf125

The piecewise constant diffusivity D(x) is given by

$$D(x) = \begin{cases} D_1, & x < \ell, \\ D_2, & x > \ell, \end{cases}$$
(A3)

so we can solve the problem separately in the two regions,

$$\frac{\partial P}{\partial t} = \begin{cases} D_1 \frac{\partial^2 P}{\partial x^2}, & -\infty < x < \ell, \\ D_2 \frac{\partial^2 P}{\partial x^2}, & \ell < x < \infty, \end{cases}$$
(A4)

and properly match the solutions at $x = \ell$. Let us introduce the functions f(x,t) and g(x,t) as follows:

$$P(x,t) = \begin{cases} f(x,t), & x < \ell, \\ g(x,t), & x > \ell. \end{cases}$$
(A5)

The continuity of the probability density P(x,t) and of the probability flux $-D(x)\frac{\partial P(x,t)}{\partial x}$ at $x = \ell$ yield two conditions:

$$f(\ell,t) = g(\ell,t) , \quad D_1 \frac{\partial f}{\partial x}\Big|_{x=l} = D_2 \frac{\partial g}{\partial x}\Big|_{x=l}.$$
(A6)

The problem can be solved by Laplace transform. Alternatively, one can look for the solutions in the two regions as sums of two rescaled and shifted Gaussians. As one can check explicitly, the resulting solution (which is not new, see e.g. [33]) is the following:

$$P(x,t) = \begin{cases} \frac{1}{\sqrt{4\pi D_{1}t}} \left[e^{-\frac{x^{2}}{4D_{1}t}} + \frac{\sqrt{D_{1}} - \sqrt{D_{2}}}{\sqrt{D_{1}} + \sqrt{D_{2}}} e^{-\frac{(x-2l)^{2}}{4D_{1}t}} \right], & x < \ell, \\ \frac{1}{\sqrt{\pi t} \left(\sqrt{D_{1}} + \sqrt{D_{2}} \right)} e^{-\frac{\left[x-\ell \left(1 - \sqrt{\frac{D_{2}}{D_{1}}}\right)\right]^{2}}{4D_{2}t}}, & x > \ell. \end{cases}$$
(A7)

In section 2 we compare the exact result (A7) for $x > \ell$ with our prediction from geometrical optics, see equation (11).

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